

<b>1.</b>	<b>Course title:</b>  Statistical Inference
<b>2.</b>	<b>Lecturer:</b>  dr hab. Radosław Poleski, rpoleski@astrouw.edu.pl
<b>3.</b>	<b>Field, type and level of studies, year of study:</b>  statistics, all years of study
<b>4.</b>	<b>Course character:</b>  General interdisciplinary lecture
<b>5.</b>	<b>Teaching method:</b>  In-person
<b>6.</b>	<b>Language:</b> English
<b>7.</b>	<b>Course type and number of hours:</b>  Workshop, 3 lessons (each 45 min.) per week
<b>8.</b>	<b>Estimated load of student's independent work:</b> 15h
<b>9.</b>	<b>Total workload and number of ECTS points:</b> 60 h, 3 ECTS
<b>10.</b>	<b>Short description and main focus of the course:</b>  The goal of the course is to prepare participants for conducting statistical inference using modern tools. The main topics of the course: <ul style="list-style-type: none"> <li>• Monte-Carlo integration</li> <li>• importance sampling</li> <li>• Markov chain Monte-Carlo</li> <li>• nested sampling</li> <li>• hierarchical Bayesian statistics</li> <li>• Bayesian model selection</li> <li>• mixture models</li> <li>• Gaussian processes</li> </ul> Topics will be adjusted based on the participants interests.
<b>11.</b>	<b>References:</b> <ol style="list-style-type: none"> <li>1. Hogg, Myers, and Bovy 2010 „Inferring the Eccentricity Distribution”, <a href="#">ApJ 725, 2166</a></li> <li>2. Foreman-Mackey, Hogg, and Morton 2014 „Exoplanet Population Inference and the Abundance of Earth Analogs from Noisy, Incomplete Catalogs”, <a href="#">ApJ 795, 64</a></li> <li>3. Sharma 2017 „Markov Chain Monte Carlo Methods for Bayesian Data Analysis in Astronomy”, <a href="#">ARA&amp;A 55, 213</a></li> <li>4. Ivezić et al. 2019 „Statistics, Data Mining, and Machine Learning in Astronomy: A Practical Python Guide for the Analysis of Survey Data”, Princeton University Press</li> </ol>

	<p>5. Speagle 2019 „A Conceptual Introduction to Markov Chain Monte Carlo Methods”, <a href="https://arxiv.org/abs/1909.12313">arXiv:1909.12313</a></p> <p>6. Aigrain and Foreman-Mackey 2023 „Gaussian Process Regression for Astronomical Time Series”, <a href="https://araa.oxfordjournals.org/advance-article-abstract/61/3/329">ARA&amp;A 61, 329</a></p>	
12.	<p><b>Prerequisites:</b></p> <ul style="list-style-type: none"> <li>• probability theory</li> <li>• programming and data visualization - python is suggested but other languages are allowed</li> </ul>	
13.	<p><b>Educational outcomes:</b></p> <p><b>Knowledge:</b> Upon successful completion of the course, the learner will demonstrate advanced knowledge and understanding of:</p> <ul style="list-style-type: none"> <li>• Theoretical foundations and practical implications of Monte-Carlo integration and variance-reduction techniques, including importance sampling.</li> <li>• Principles, mathematics, and assumptions underpinning Markov Chain Monte Carlo, nested sampling, and hierarchical Bayesian models.</li> <li>• Bayesian model selection criteria, their interpretation, strengths and limitations in complex inference problems.</li> <li>• The structure and properties of mixture models and the role of Gaussian processes in modelling uncertainty and functions.</li> <li>• State-of-the-art literature and research developments in Bayesian statistics, including conceptual challenges and areas of active methodological research.</li> </ul> <p><b>Practical Skills:</b> Upon successful completion of the course, the learner will be able to:</p> <ul style="list-style-type: none"> <li>• Independently design and implement Monte Carlo and MCMC algorithms using modern computational tools to perform efficient statistical inference.</li> <li>• Apply nested sampling and importance sampling methods to estimate posterior distributions and model evidence in complex statistical models.</li> <li>• Develop and critically evaluate hierarchical Bayesian models.</li> <li>• Select and compare competing models using Bayesian model selection techniques and interpret the results in context.</li> <li>• Employ mixture models and Gaussian processes in data analysis, including implementation, tuning, and</li> </ul>	<p><b>PQF level 8 codes:</b></p> <p>P8S_WG</p> <p>P8S_UW, P8S_UO</p>

	<p>validation in programmes or software environments appropriate for advanced statistical computing.</p> <p><b>Social Skills:</b> Upon successful completion of the course, the learner will be ready to:</p> <ul style="list-style-type: none"> <li>• Critically reflect on the ethical, theoretical and practical implications of statistical modelling choices, including responsibility for transparent and reproducible analysis.</li> <li>• Communicate complex statistical concepts, results and their limitations effectively to both specialist and non-specialist audiences.</li> <li>• Work autonomously and collaboratively in research or professional teams to integrate advanced inference methods into interdisciplinary projects.</li> <li>• Engage in independent study and continuous professional development, synthesising new research findings with existing knowledge to drive innovation in statistical practice.</li> </ul>	P8S_KK, P8S_KR
14.	<b>Evaluation of the educational outcomes:</b> homework assignments and oral exam	
15.	<b>Criteria to complete the course:</b> Active attendance – laptop for coding is required; homework assignments in the case of more than two classes missed.	
16.	<b>Contact with the lecturer:</b> email: <a href="mailto:rpoleski@astrouw.edu.pl">rpoleski@astrouw.edu.pl</a> in-person meeting (at Al. Ujazdowskie 4, room 3080) possible after making an appointment	